

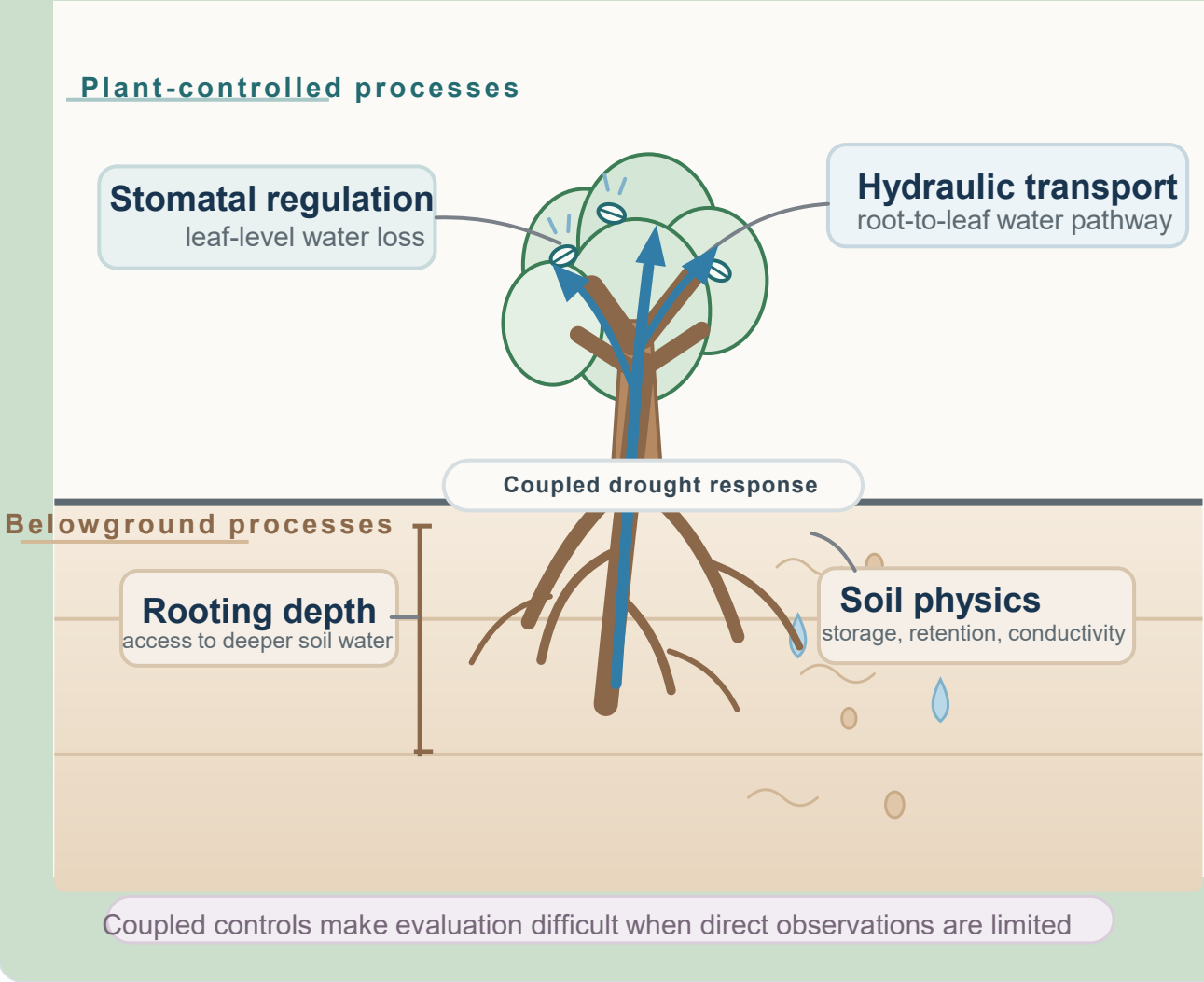
Evaluating drought responses in JULES at FR-Pue

*From structural calibration and stomatal optimisation
To model transferability and hydraulic legacy*

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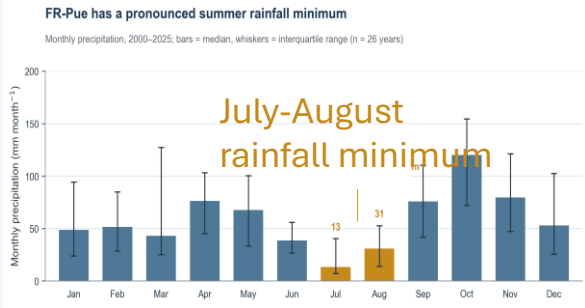
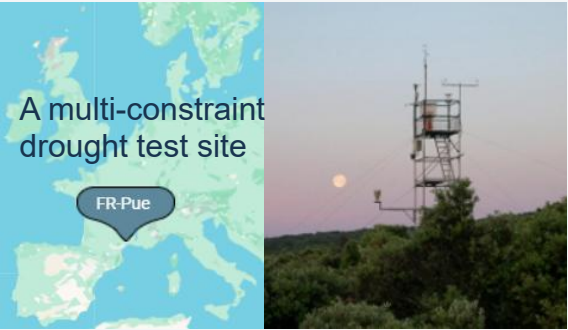
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Background



1 | Observed data FR-Pue / Puéchabon

Mediterranean oak forest
Southern France
Recurrent seasonal drought



Flux (2000-2025)

- GPP
- raw LE
- LEcorr

Water statu

- Ψ_{leaf}
- Soil water content

2 | Models used

JULES Baseline

- **STD-OOB**
(Standard JULES: out-of-box baseline)
- **PM-OOB**
(Profit-max stomatal optimisation out-of-box baseline)

Soil depth and layering
Soil parameters: b, sathh, sm_sat
Photosynthetic : vsI
Root zone water potential: root_psi_crit

Control soil structural and parameters

- **STD-TUNED**
- **PM-TUNED**

New hydraulic legacy model

- **PM-XI (Xylem impairment)**

Research questions

1

How much can model bias be reduced?

How much can bias be reduced by constraining belowground structure and key parameters?

2

Does stomatal optimisation add value beyond calibration?

Does Profit-Max improve predictions of drought fluxes beyond calibration?

3

Are improvements for the right reasons?

Do improvements arise from more realistic physiological responses?

4

Can the calibrated model predict a more severe drought?

Can JULES simulate a more intense rain-exclusion experiment it was not calibrated for?

5

Can hydraulic legacy effects improve drought predictions?

How do previous summers influence ecosystem responses in current summer?

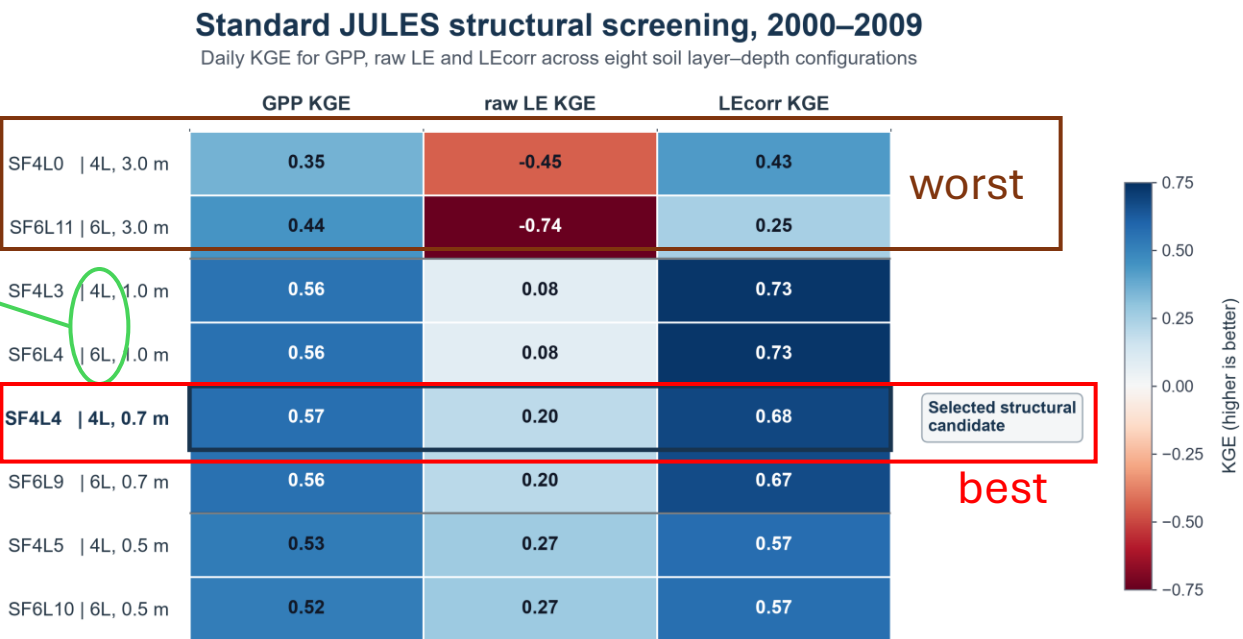
1

**How much of the bias can be reduced
by constraining belowground
structure and key parameters?**

Result: Soil structure calibration

STD 8 soil layer–depth configurations | daily GPP(GPP-prioritised), raw LE and LECorr | 2000–2009 calibration period

1 | Structural screening



2 | Summary

Main pattern

- 3 m: perform worst
- 0.7 m: higher overall skill
- 4 vs 6 layers: small differences within same depth

Result

- Constraining soil-column depth removes much of the baseline bias;
- layer-number effects are comparatively small.

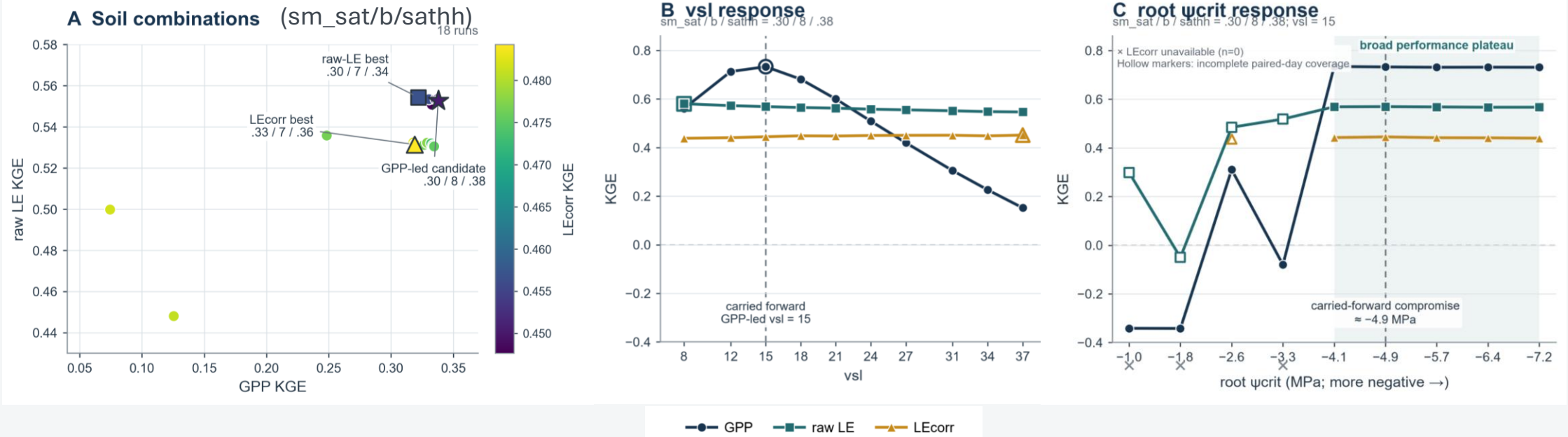
Selected structural candidate

GPP-prioritised

- STD: SF4L4 (4 layers, 0.7 m)
- PM: SF6L10 (6 layers, 0.5 m)

Result: soil, photosynthesis, hydraulic parameters

PM soil, vsl and root ψ crit | daily GPP, raw LE and LEcorr | 2000–2009 calibration period



GPP-best point, the raw-LE best point and the LEcorr best point are not same

GPP-prioritised selected:
sm_sat/b/sathh = 0.3/8/0.38

GPP is sensitive to vsl and peaks around 15

GPP-prioritised selected:
vsl = 15

Below roughly -4 MPa the metrics enter a broad plateau

GPP-prioritised selected:
root_psi_crit = -4.9

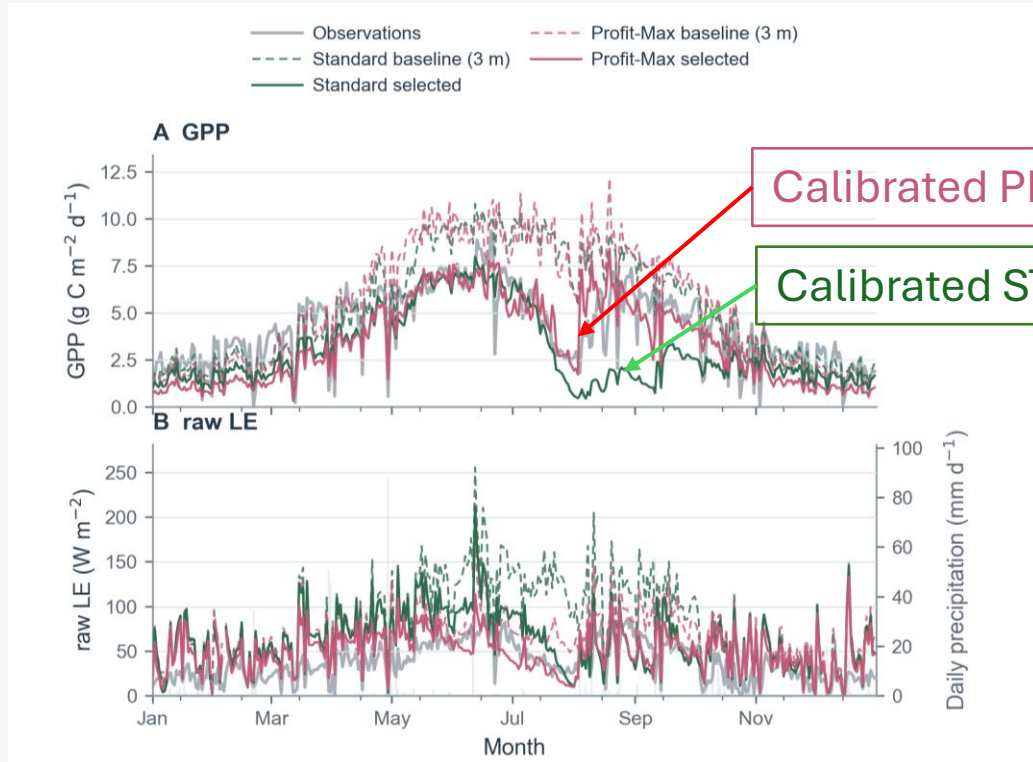
Summary: No single optimum across GPP, raw LE and LEcorr. Calibration target controls the selected parameters.

2 Does stomatal optimisation improve drought fluxes beyond calibration?

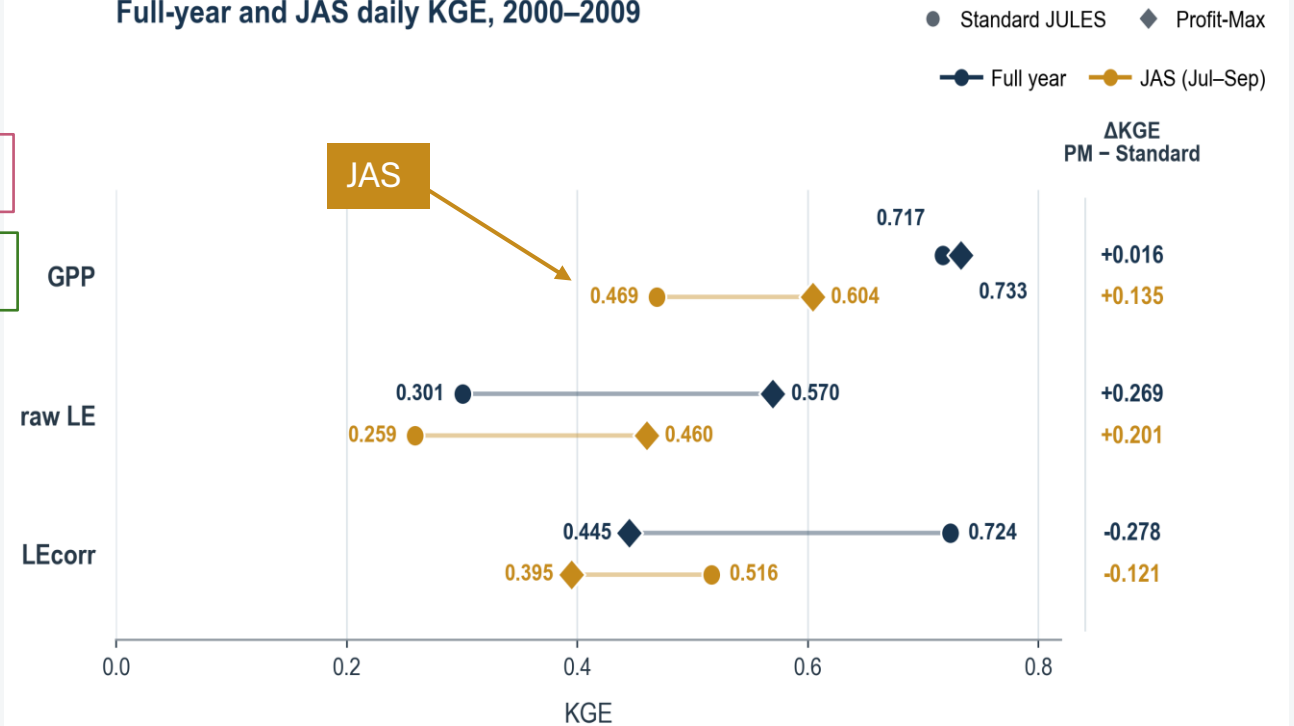
Result: Comparison between tuned STD and PM

Full-year and JAS | 2000–2009 | GPP, raw LE and LEcorr

2004 GPP and LE



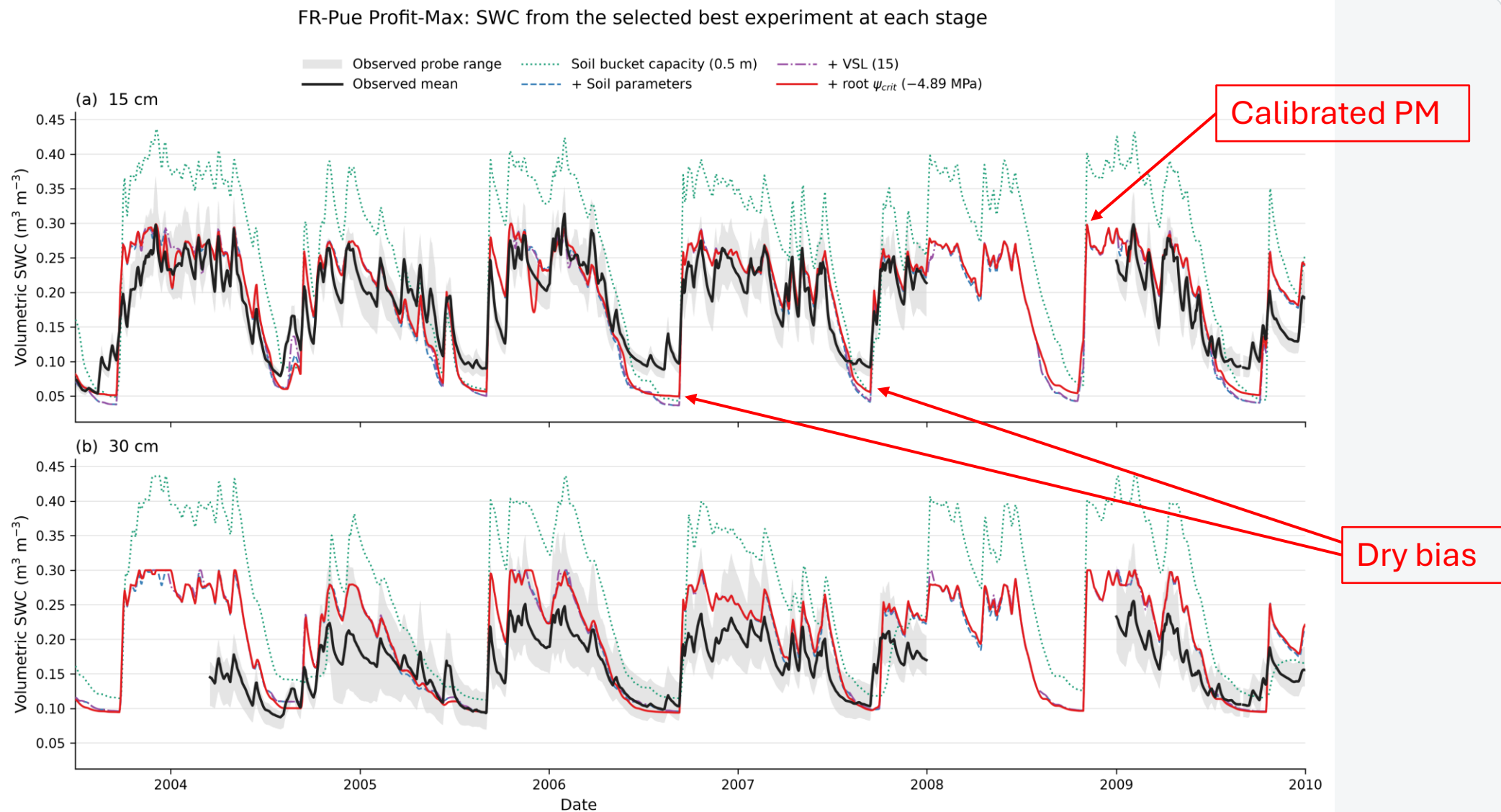
Full-year and JAS daily KGE, 2000–2009



Summary: Profit-Max shows a GPP advantage in summer. Improving plant hydrology adds value on top of calibration.

3 Are improvements achieved for the right physiological reasons?

Result: Evaluation of simulated soil water content

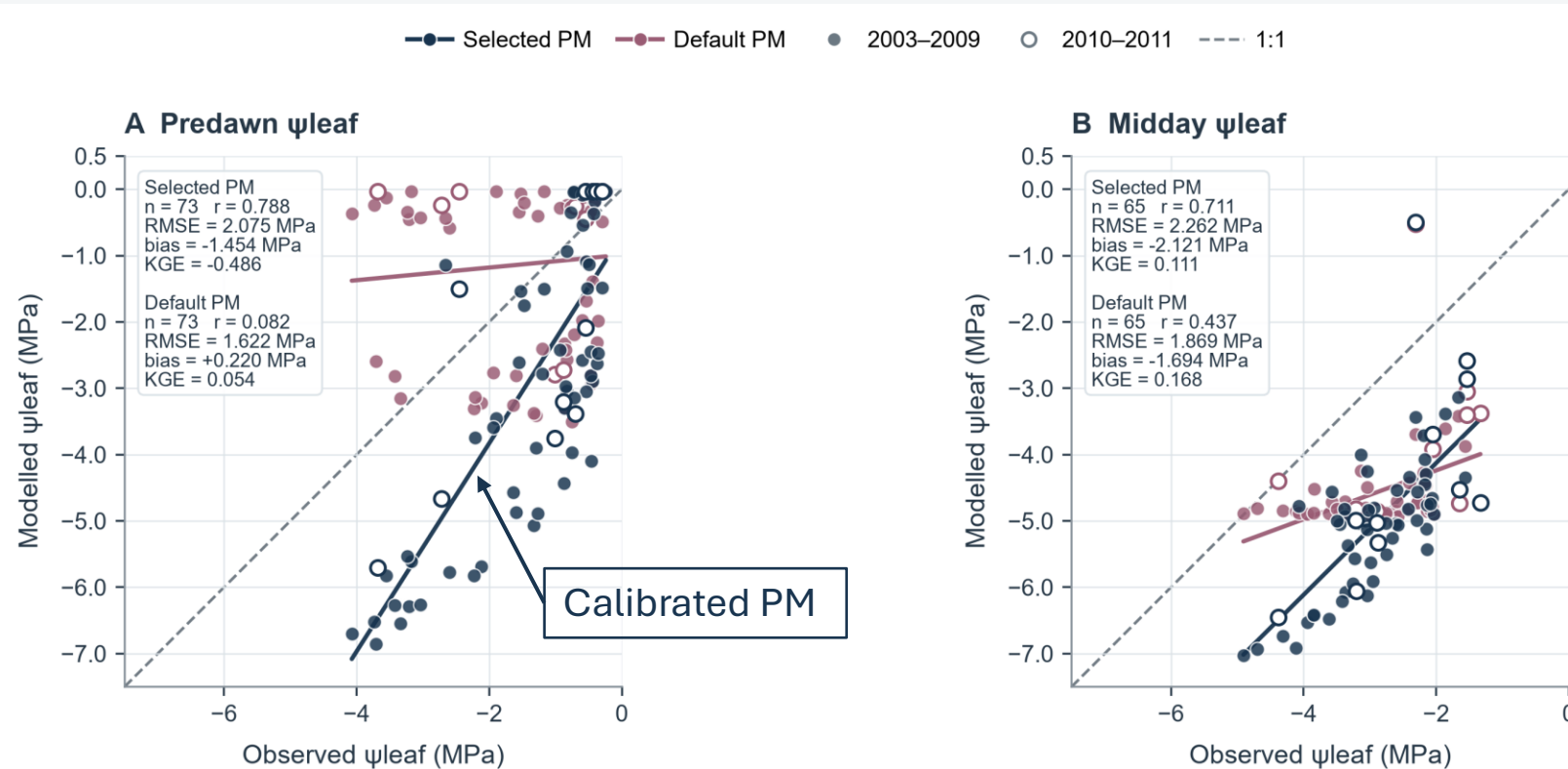


Summary: PM captures the SWC dynamics well, with a slight dry bias at 15 cm.

Result: Leaf-water-potential improvement

more negative:
stronger water stress

Selected versus default Profit-Max | historical ψ_{leaf} observations | 2003–2011



What improved?

- Predawn correlation improves strongly
 $r: 0.082 \rightarrow 0.788$
- Midday correlation also improves
 $r: 0.437 \rightarrow 0.711$

What did not improve?

- Predawn RMSE increases
 $1.622 \rightarrow 2.075$ MPa
- Midday RMSE increases
 $1.869 \rightarrow 2.262$ MPa
- **Bias becomes more negative**
predawn $+0.220 \rightarrow -1.454$ MPa
midday $-1.694 \rightarrow -2.121$ MPa

Summary

Calibrated PM improves ψ_{leaf} correlation, but stress magnitude is too strong

H1: LAI bias $\rightarrow$ too much water being used? Rejected
H2: Esoil bias $\rightarrow$ plant doesn't have enough water? Some evidence

4

**Can the calibrated model predict a
more severe drought?**

Result: Leaf-water-potential in rain-exclusion treatment



Leading into
drought



More intense
drought period



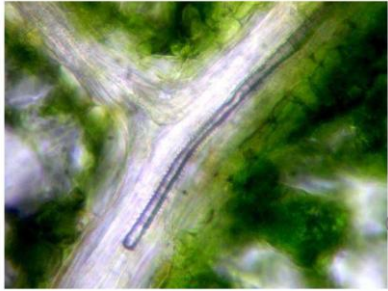
Recovery from
drought



Summary: When interpreting the treatment results, we should keep in mind that the model already tends to simulate leaf water potential values that are too negative.

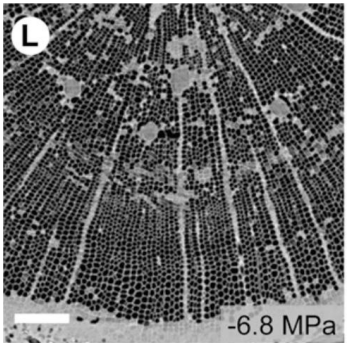
5 Can hydraulic legacy effects improve drought predictions?

Result: New hydraulic legacy model with xylem impairment



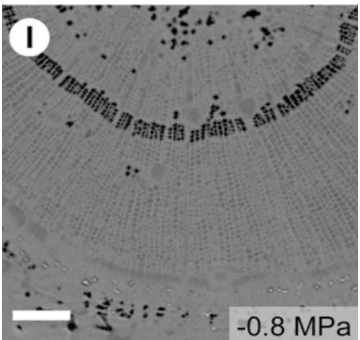
Air blockage
("embolism") due
to water stress

©INRA/Hervé Cochard

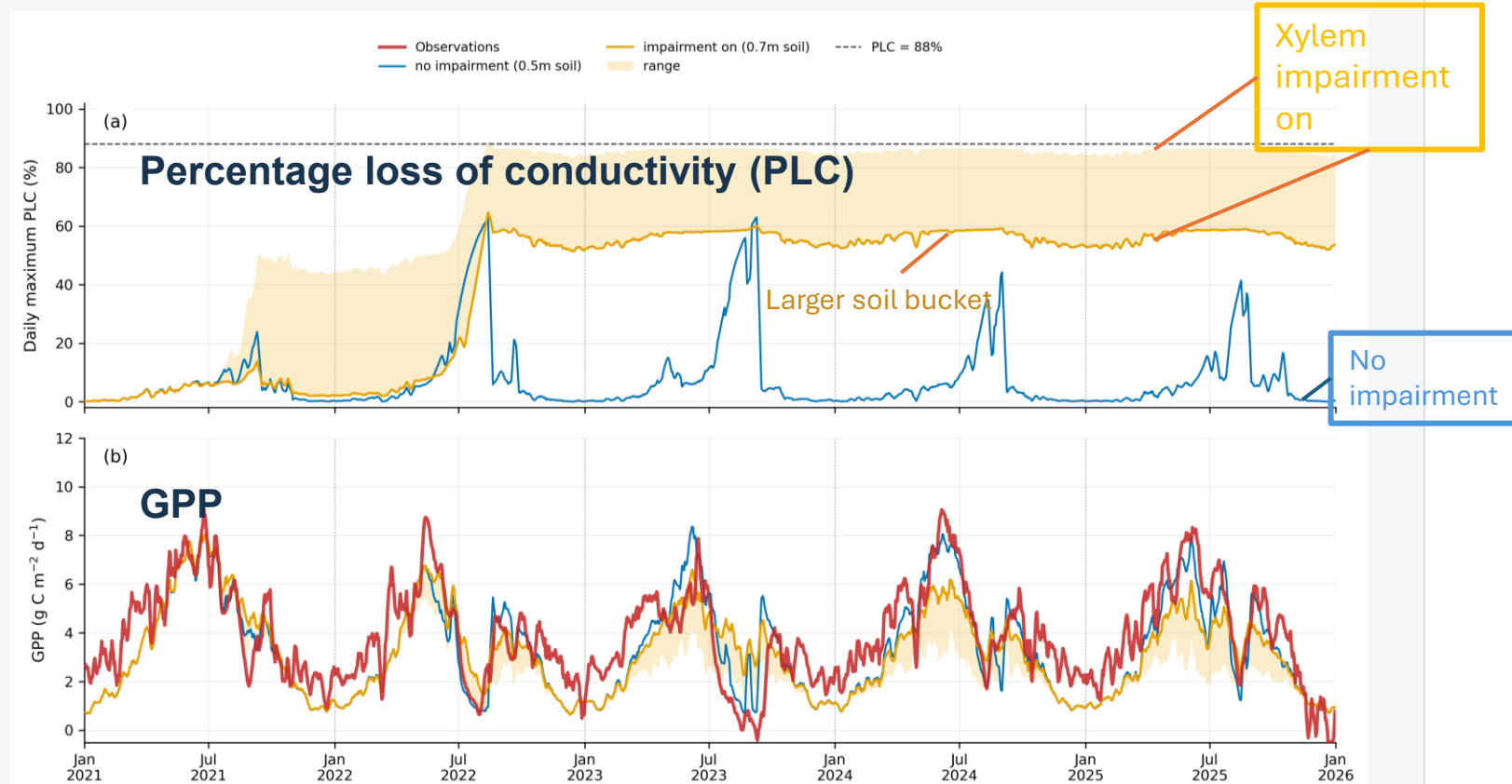


Spread of
embolism
(gas bubbles that
block xylem
conduits)

Choat et al. (2016)
Plant Phys.



Branch
dieback,
even death



Summary:

- Simulated PLC is highly sensitive to soil-bucket depth (a larger soil bucket reduces xylem damage) although parameter uncertainty remains.
- The model overestimates legacy damage because it does not regrowth of new xylem.

Summary

1

Structural calibration removes a major part of the baseline bias,
but no unique parameter optimum emerges – linked to LE bias

2

Profit-Max improves summer GPP and raw LE,
but the advantage is not reproduced for LEcorr.

3

SWC and ψ_{leaf} dynamics are broadly well represented
but ψ_{leaf} stress magnitude remains too strong.

4

Transferability to intensified drought remains preliminary
and baseline-bias dependent.

5

Hydraulic legacy can strongly amplify consecutive drought effects,
but the current implementation lacks the capacity to recovery (needs dgvm)

Thanks for listening. — Questions?

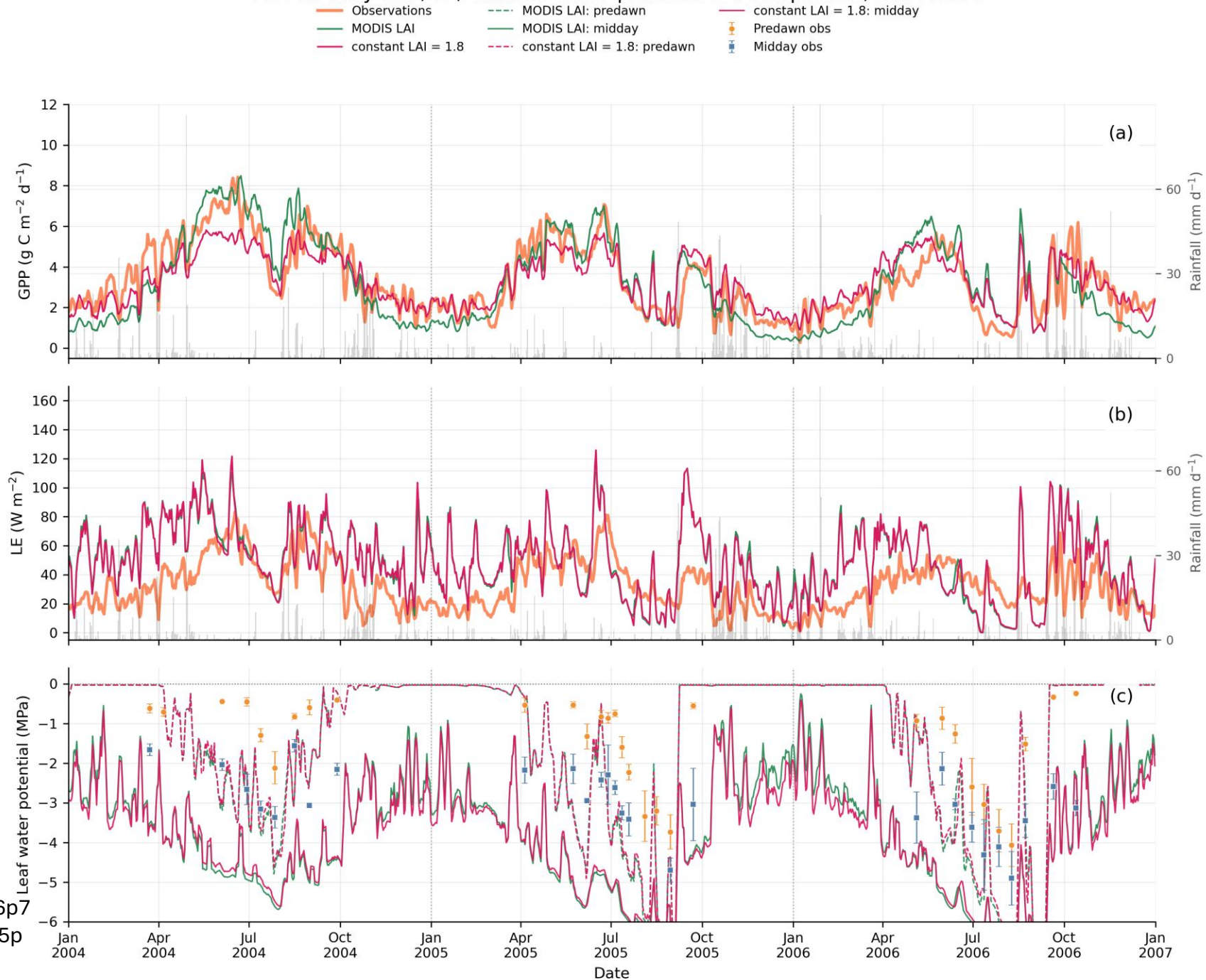


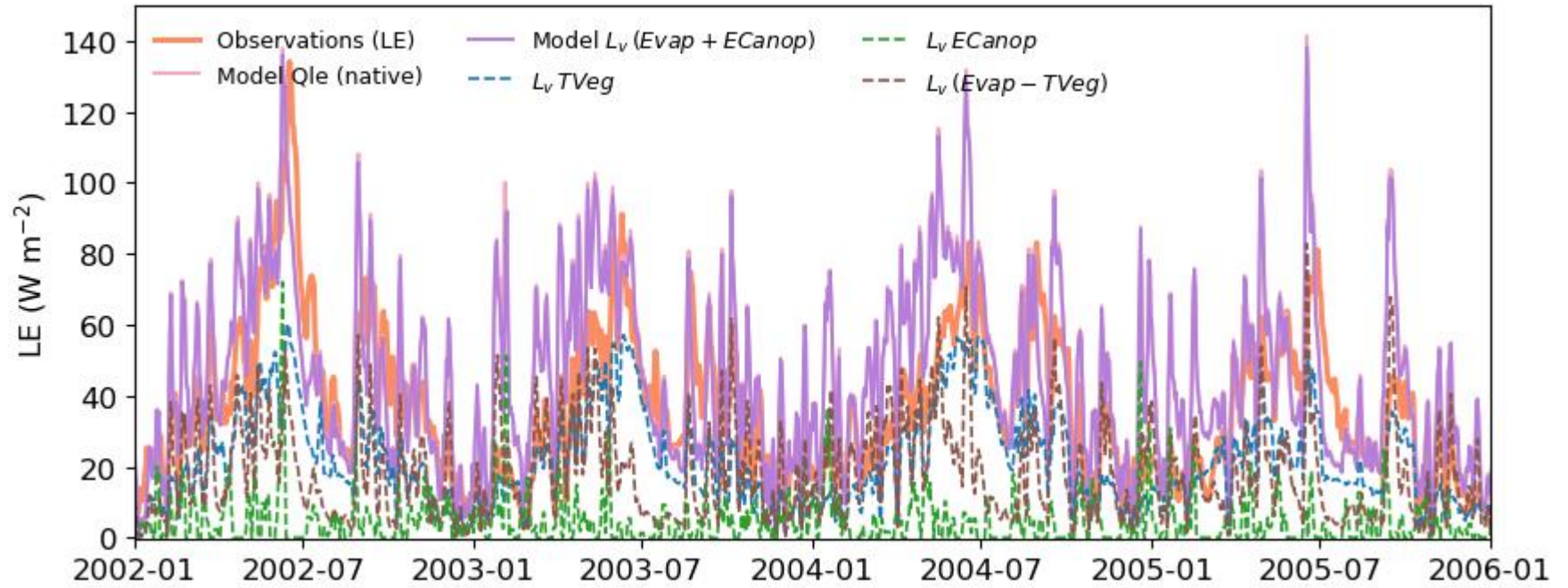
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FR-Pue daily GPP, LE, and leaf water potential LAI comparison, 2004-2006





Bias: model LE – OBS LE

[corr] bias~TVeg : $r = 0.360$, $p = 3.457e-79$

[corr] bias ~ECanop : $r = 0.381$, $p = 4.422e-89$

[corr] bias ~ESoil : $r = 0.654$, $p = 1.986e-312$

